

Final

- Dec. 8, 2-5pm, GB304
- Closed book
- Entire course
- Proofs: Chapter 3 and 5

1

Renewal Reward Process

- If $\bar{X} < \infty$ or $E[R_n] < \infty$, then

$$\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t R(\tau) d\tau = \frac{E[R_n]}{\bar{X}}, \quad \text{with probability 1.}$$

- If non-arithmetic renewal process and $r(z)$ is directly Riemann integrable, then

$$\lim_{t \rightarrow \infty} E[R(t)] = \frac{E[R_n]}{\bar{X}}.$$

- If arithmetic renewal process with span d , then

$$\lim_{n \rightarrow \infty} E[R(nd)] = \frac{E[R_n]}{\bar{X}}.$$

3

Renewal Process

Assumptions: X_1, X_2, \dots positive IID r.v.; $\bar{X} < \infty$,

Results: (note that $0 < \bar{X}$)

$$\lim_{t \rightarrow \infty} \frac{N(t)}{t} = \frac{1}{\bar{X}}, \quad w.p.1$$

$$\lim_{t \rightarrow \infty} \frac{E[N(t)]}{t} = \frac{1}{\bar{X}}$$

$$\lim_{t \rightarrow \infty} (E[N(t + \delta)] - E[N(t)]) = \frac{\delta}{\bar{X}}, \quad \delta > 0; \text{ (non-arithmetic)}$$

2

Finite State Markov Chain

- Questions

- Does $\pi = \pi[P]$ have a probability vector solution?
- Does $\pi = \pi[P]$ have a unique probability vector solution?

$$\text{- Is } \lim_{n \rightarrow \infty} [P]^n = \begin{bmatrix} \pi_1 & \cdots & \pi_J \\ \vdots & & \vdots \\ \pi_1 & \cdots & \pi_J \end{bmatrix}$$

4

Finite State Markov Chain - Answers

- Yes, solution to $\pi = \pi[P]$ always exists
- Unique, if and only if there is a single recurrent class (and possibly many transient classes)
- If there are r recurrent classes, then there exist r linearly independent solutions.
- For ergodic Markov chain we have

$$\lim_{n \rightarrow \infty} [P]^n = \begin{bmatrix} \pi_1 & \cdots & \pi_J \\ \vdots & & \vdots \\ \pi_1 & \cdots & \pi_J \end{bmatrix}$$

- If there are several multiple recurrent classes $\lim_{n \rightarrow \infty} [P]^n$ exists, but rows are not identical.
- If there is one or more periodic class then $[P]^n$ does not converge.

5

Markov Chains with Countably Infinite State Spaces

- T_{ij} : "first passage time from i to j "
 - $f_{ij}(n)$: probability mass function
 - $F_{ij}(n)$: probability distribution function
- Classification
 - **recurrent**: state i is recurrent if $F_{ii}(\infty) = 1$
 - * **positive recurrent** $E[T_{ii}] < \infty$
 - * **null recurrent** $E[T_{ii}] = \infty$
 - **transient**: state i is transient if $F_{ii}(\infty) < 1$

7

Finite State Markov Chain with Rewards

- Single Recurrent Class: $v(n) = nge + w + [P]^n\{v(0) - w\}$
- Ergodic: $\lim_{n \rightarrow \infty} \{v(n) - nge\} = w + \beta e$, $\beta = \pi(v(0) - w)$

6

Renewal Theory

Assume that state j is recurrent and consider the renewal process $\{N_{jj}(t); t \geq 0\}$. Then

- $\lim_{t \rightarrow \infty} \frac{N_{jj}(t)}{t} = \frac{1}{E[T_{jj}]}$, *w.p.1*
- $\lim_{t \rightarrow \infty} \frac{E[N_{jj}(t)]}{t} = \frac{1}{E[T_{jj}]}$
- If state j is periodic with span d , then

$$\lim_{n \rightarrow \infty} P(X_{nd} = j \mid X_0 = j) = \frac{d}{T_{jj}}$$

- If state j is aperiodic, then

$$\lim_{n \rightarrow \infty} P(X_n = j \mid X_0 = j) = \frac{1}{T_{jj}}$$

8

Steady-State Distribution

Vector $(\pi_0, \pi_1, \pi_2, \dots)$, $\pi_i \geq 0$, such that

$$\pi_i = \sum_j \pi_j P_{ji}$$

and

$$\sum_i \pi_i = 1$$

- **Theorem:** Consider an irreducible M.C. with transition probabilities $\{P_{ij}\}$. If the above equation has a solution, the the solution is **unique**, we have $\pi_i = \frac{1}{T_{ii}}$ for all $i \geq 0$, and all states are **positive recurrent**. Also, if all states are positive recurrent, then the above equation has a solution.

9

Semi-Markov Processes

- **Lemma:** Consider a semi-Markov process with an **irreducible** recurrent embedded M.C. $\{X_n : n \geq 0\}$. Given $X_0 = i$, let $M_{ij}(t)$ be the number of transitions into a given state j in the interval $(0, t]$. Then $M_{ij}(t)$ is a delayed renewal process (and $M_{jj}(t)$ is a renewal process).

- **Theorem:** Assume that the embedded M.C. of a semi-Markov process is irreducible and positive recurrent. If $\sum_i \pi_i \bar{U}(i) < \infty$, then, with probability 1, the limiting fraction of time spent in state i is

$$p_i = \frac{\pi_i \bar{U}(i)}{\sum_j \pi_j \bar{U}(j)}.$$

11

Semi-Markov Processes

- $X(t)$ state of process at time $t \geq 0$, $X(t) \in \{0, 1, 2, \dots\}$
- $S_1 < S_2 < S_3 < \dots$: epochs at which transitions occur
- X_n : new state entered at time n : $X_n = X(S_n)$, $X(t) = X_n$ for $S_n \leq t < S_{n+1}$
- $\{X_n : n \geq 0\}$ is a M.C. with $\{P_{ij}\}$. This M.C. is called the embedded M.C.
- $U_n = S_n - S_{n-1}$ is a R.V
 - depends only on X_{n-1}, X_n

$$P(U_n \leq u \mid X_{n-1} = i, X_n = j) = G_{ij}(u)$$

- $\bar{U}(i, j)$: conditional mean of transition time
- $\bar{U}(i) = \sum P_{ij} \bar{U}(i, j)$

10

Special Cases

- Markov chains with countably infinite state spaces
 - Branching processes
 - Birth-death Markov chains
 - Reversible Markov chains
- Semi-Markov processes
 - Markov processes
 - * Birth-Death Markov processes
 - * Reversible Markov

12

Dyanmic Programming

- Extension of finite state Markov chains with rewards

- Optimal stationary policy B

- Single Recurrent Class:

$$v^B(n) = ng^B e + w^B + [P^B]^n \{v^B(0) - w^B\}$$

- Ergodic:

$$\lim_{n \rightarrow \infty} \{v^B(n) - ng^B e\} = w^B + \beta e; \quad \beta = \pi^B(v(0) - w^B)$$